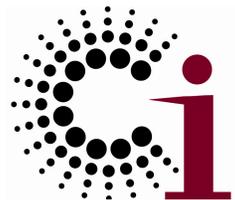




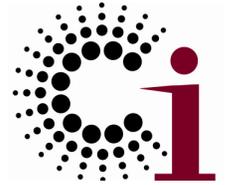
Fast, Reliable, Loosely Coupled Parallel Computation

Ian Foster

Computation Institute
Argonne National Laboratory
University of Chicago

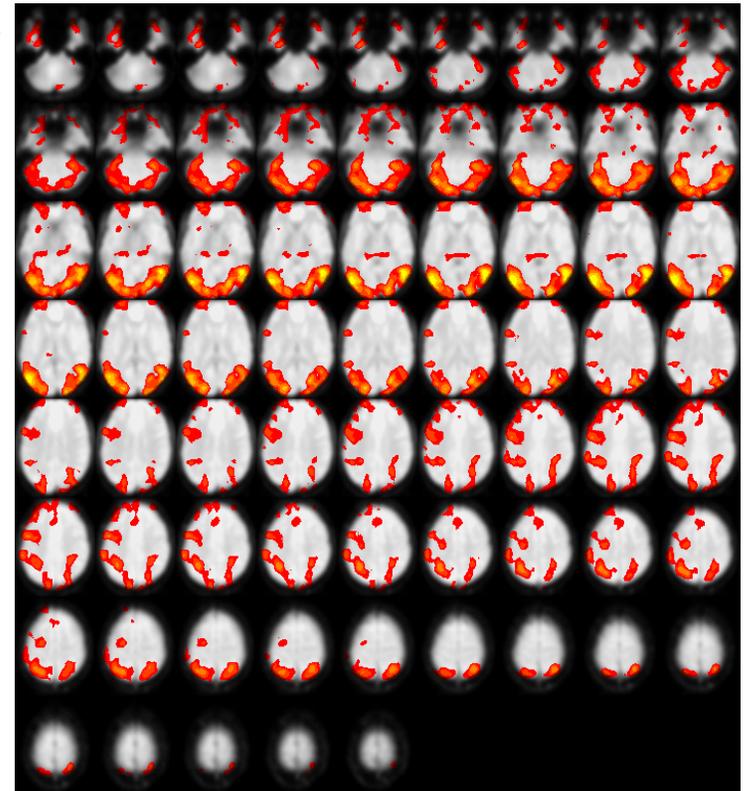


Joint work with **Yong Zhao, Ioan Raicu, Mike Wilde, Ben Clifford,
Mihael Hatigan, Tibi Stef-Praun, Veronika Nefedova**



Case Study: The Functional MRI (fMRI) Data Center

- Online repository of neuroimaging data
 - ◆ A typical study comprises 3 groups, 20 subjects/group, 5 runs/subject, 300 volumes/run
 - 90,000 volumes, 60 GB raw
 - 1.2 million files processed
 - ◆ 100s of such studies in total
- Many users analyze this data
 - ◆ Wide range of analyses
 - ◆ Testing → production
 - ◆ Ensembles: a set of data analyses by parameters, datasets

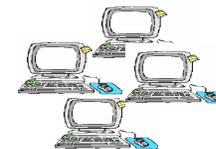
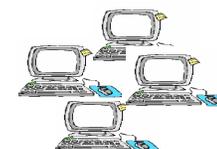
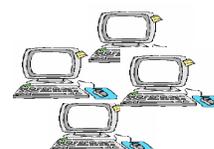
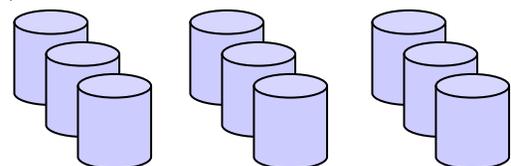
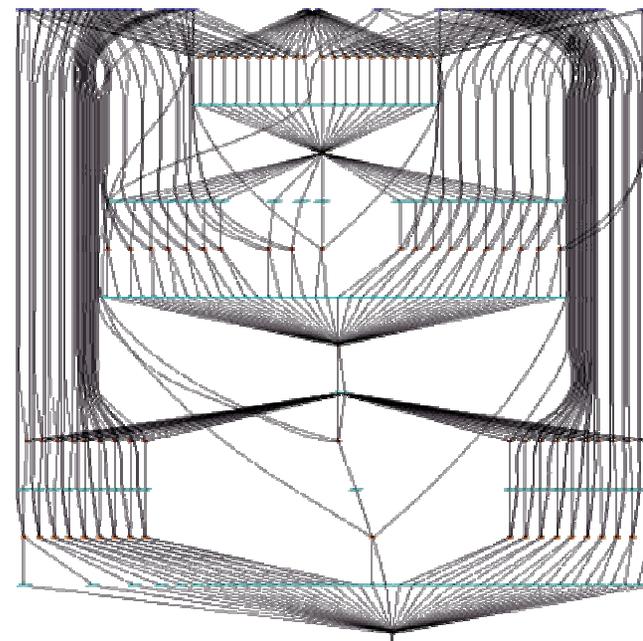
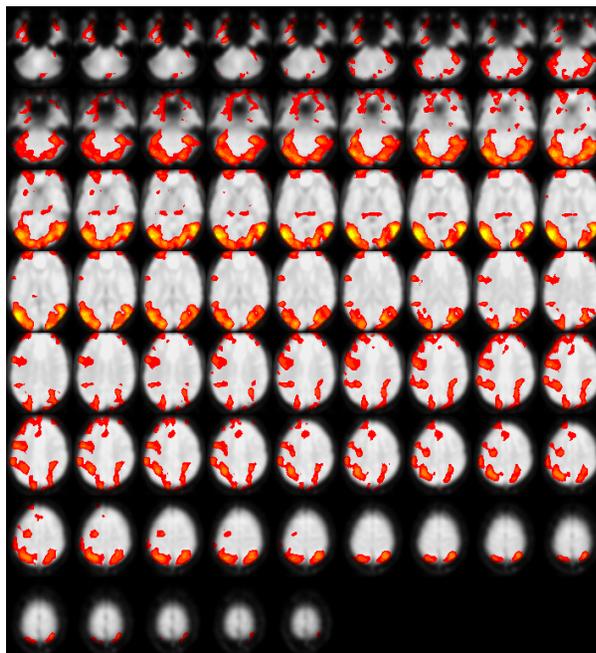
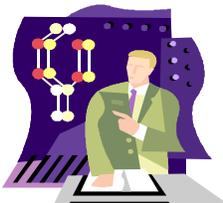
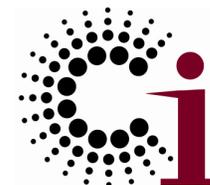




the globus alliance

www.globus.org

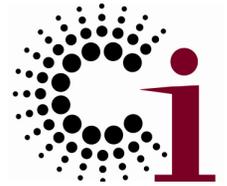
fMRI: A Broad Picture





Challenges

- Deluge of data: instrumentation, simulation
- Data analysis turns into data integration
- Community-wide collaboration
- Provenance: tracking, query, application
- Scalability: desktop to Grid
- Productivity: throughput, performance



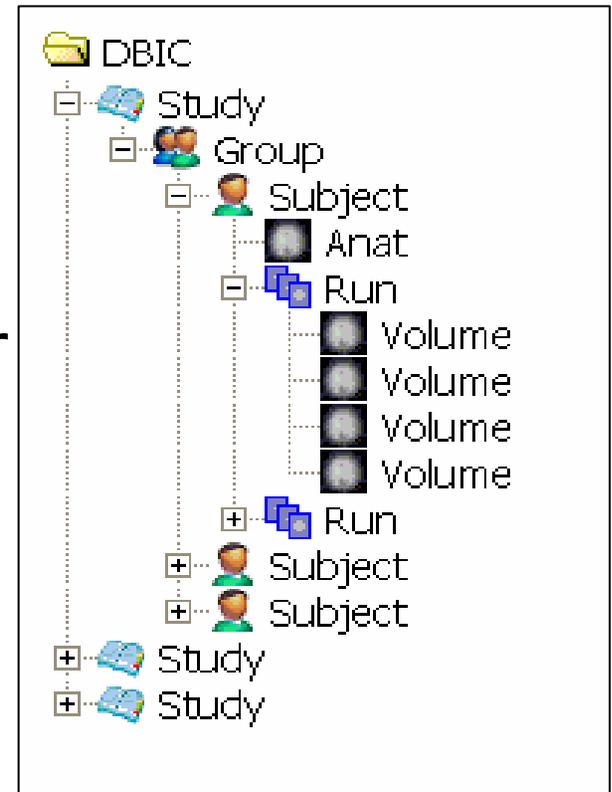
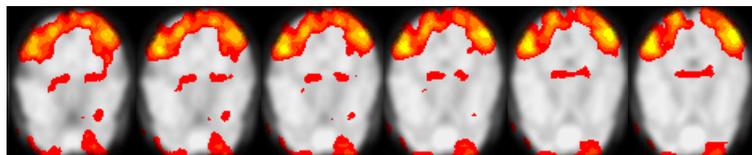
Swift System

- Clean separation of logical/physical concerns
 - ◆ **XDTM** specification of logical data structures
- + Concise specification of parallel programs
 - ◆ **SwiftScript**, with iteration, etc.
- + Efficient execution on distributed resources
 - ◆ **Karajan** threading, **Falkon** provisioning, **Globus** interfaces, pipelining, load balancing
- + Rigorous provenance tracking and query
 - ◆ Virtual data schema & automated recording
- **Improved usability and productivity**
 - ◆ Demonstrated in numerous applications



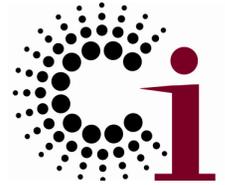
The Messy Data Problem

- Scientific data is often logically structured
 - ◆ E.g., hierarchical structure
 - ◆ Common to map functions over dataset members
 - ◆ Nested map operations can scale to millions of objects





The Messy Data Problem



- But physically “messy”
- Heterogeneous storage format and access protocol
 - ◆ Logically identical dataset can be stored in textual File (e.g. CSV), spreadsheet, database, ...
 - ◆ Data available from filesystem, DBMS, HTTP, WebDAV, ..
- Metadata encoded in directory and file names
- Hinders program development, composition, execution

```
./knottastic
total 58
drwxr-xr-x 4 yongzh users 2048 Nov 12 14:15 AA
drwxr-xr-x 4 yongzh users 2048 Nov 11 21:13 CH
drwxr-xr-x 4 yongzh users 2048 Nov 11 16:32 EC

./knottastic/AA:
total 4
drwxr-xr-x 5 yongzh users 2048 Nov 5 12:41 04nov06aa
drwxr-xr-x 4 yongzh users 2048 Dec 6 12:24 11nov06aa

./knottastic//AA/04nov06aa:
total 54
drwxr-xr-x 2 yongzh users 2048 Nov 5 12:52 ANATOMY
drwxr-xr-x 2 yongzh users 49152 Dec 5 11:40 FUNCTIONAL

./knottastic/AA/04nov06aa/ANATOMY:
total 58500
-rw-r--r-- 1 yongzh users 348 Nov 5 12:29 coplanar.hdr
-rw-r--r-- 1 yongzh users 16777216 Nov 5 12:29 coplanar.img

./knottastic/AA/04nov06aa/FUNCTIONAL:
total 196739
-rw-r--r-- 1 yongzh users 348 Nov 5 12:32 bold1_0001.hdr
-rw-r--r-- 1 yongzh users 409600 Nov 5 12:32 bold1_0001.img
-rw-r--r-- 1 yongzh users 348 Nov 5 12:32 bold1_0002.hdr
-rw-r--r-- 1 yongzh users 409600 Nov 5 12:32 bold1_0002.img
-rw-r--r-- 1 yongzh users 496 Nov 15 20:44 bold1_0002.mat
-rw-r--r-- 1 yongzh users 348 Nov 5 12:32 bold1_0003.hdr
-rw-r--r-- 1 yongzh users 409600 Nov 5 12:32 bold1_0003.img
```

XML Dataset Typing & Mapping (XDTM)

- Describe logical structure by **XML Schema**
 - ◆ Primitive scalar types: int, float, string, date, ...
 - ◆ Complex types (structs and arrays)
- Use **mapping descriptors** for mappings
 - ◆ How dataset elements are mapped to physical representations
 - ◆ External parameters (e. g. location)
- Use **XPath** for dataset selection



XDTM: Related Work

- Data format standardization
 - ◆ FITS, CDF, HDF-5, DICOM
- Data format description
 - ◆ DFDL [Beckerle,Westhead04] embeds annotations with XML Schema
 - ◆ PADS [Fisher,Gruber05], PADX [Fernandez,Fisher06], declarative specs of physical layout and semantic properties
- Logical object
 - ◆ ADO [Microsoft01], in memory relational model
 - ◆ SDO [Beatty,Brodsky03], logical data model for J2EE programming



XDTM: Implementation

- Virtual integration
 - ◆ Each data source treated as virtual XML source
 - ◆ Data structure defined as XML schema
 - ◆ Mapper responsible for accessing source and translating to/from XML representation
 - ◆ Bi-directional
- Common mapping interface
 - ◆ Data providers implement the interface
 - Responsible for data access details
 - ◆ Standard mapper implementations provided
 - String, file system, CSV, ...



SwiftScript

- **Typed parallel programm** [SIGMOD05, Springer06]
 - ◆ XDTM as data model and type system
 - ◆ Typed dataset and procedure definitions
- **Scripting language**
 - ◆ Implicit data parallelism
 - ◆ Program composition from procedures
 - ◆ Control constructs (foreach, if, while, ...)

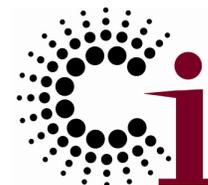
Clean application logic
Type checking
Dataset selection, iteration
Discovery by types
Type conversion

A Notation & System for Expressing and Executing Cleanly Typed Workflows on Messy Scientific Data [SIGMOD05]



SwiftScript: Related Work

- Coordination language
 - ◆ Linda[Ahuja,Carriero86], Strand[Foster,Taylor90], PCN[Foster92]
 - ◆ Durra[Barbacci,Wing86], MANIFOLD[Papadopoulos98]
 - ◆ Components programmed in specific language (C, FORTRAN) and linked with system
- “Workflow” languages and systems
 - ◆ Taverna[Oinn,Addis04], Kepler[Ludäscher,Altintas05], Triana [Churches,Gombas05], Vistrail[Callahan,Freire06], DAGMan, Star-P
 - ◆ XPDL[WfMC02], BPEL[Andrews,Curbera03], and BPML[BPML02], YAWL[van de Aalst,Hofstede05], Windows Workflow Foundation [Microsoft05]



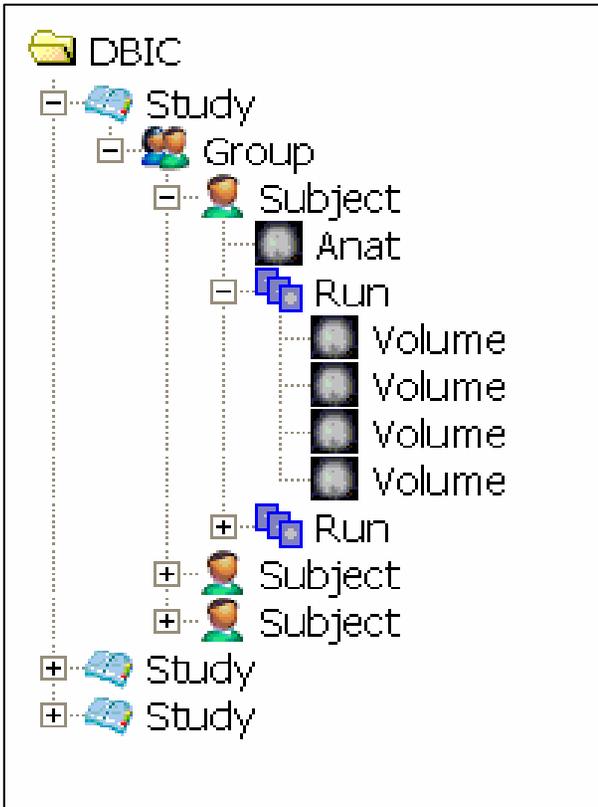
Related Work

	SwiftScript	BPEL	XPDL	MW Wflow	DAGMan	Tavana	Triana	Kepler	Vistrail	Star-P
Scales to Grids	++	-	-	-	++	-	-	-	-	+
Typing	++	++	++	++	-	-	-	+	-	+
Iteration	++	-/+	-	+	-	-	-	+	-	+
Scripting	++	-	-	+	+	+	-	-	+	++
Dataset Mapping	+	-	-	-	-	-	-	-	-	-
Service Interop	+	-	+	-	-	-	-	+	-	-
Subflow/comp.	+	-	+	+	-	-	+	+	-	+
Provenance	+	-	-	+	-	+	-	+	+	-
Open source	+	+	+	-	+	+	+	+	+	-

“A 4x200 flow leads to a 5 MB BPEL file ... chemists were not able to write in BPEL” [Emmerich,Buchart06]



fMRI Type Definitions in SwiftScript



Simplified version of
fMRI AIRSN Program
(Spatial Normalization)

```
type Study {
    Group g[ ];
}
```

```
type Group {
    Subject s[ ];
}
```

```
type Subject {
    Volume anat;
    Run run[ ];
}
```

```
type Run {
    Volume v[ ];
}
```

```
type Volume {
    Image img;
    Header hdr;
}
```

```
type Image {};
```

```
type Header {};
```

```
type Warp {};
```

```
type Air {};
```

```
type AirVec {
    Air a[ ];
}
```

```
type NormAnat {
    Volume anat;
    Warp aWarp;
    Volume nHires;
}
```



Type Definitions in XML Schema

```
<xs:schema targetNamespace="http://www.fmri.org/schema/airsn.xsd"
  xmlns="http://www.fmri.org/schema/airsn.xsd"
  xmlns:xs="http://www.w3.org/2001/XMLSchema">
  <xs:simpleType name="Image">
    <xs:restriction base="xs:string"/>
  </xs:simpleType>
  <xs:simpleType name="Header">
    <xs:restriction base="xs:string"/>
  </xs:simpleType>
  <xs:complexType name="Volume">
    <xs:sequence>
      <xs:element name="img" type="Image"/>
      <xs:element name="hdr" type="Header"/>
    </xs:sequence>
  </xs:complexType>
  <xs:complexType name="Run">
    <xs:sequence minOccurs="0" maxOccurs="unbounded">
      <xs:element name="v" type="Volume"/>
    </xs:sequence>
  </xs:complexType>
</xs:schema>
```



AIRSN Program Definition

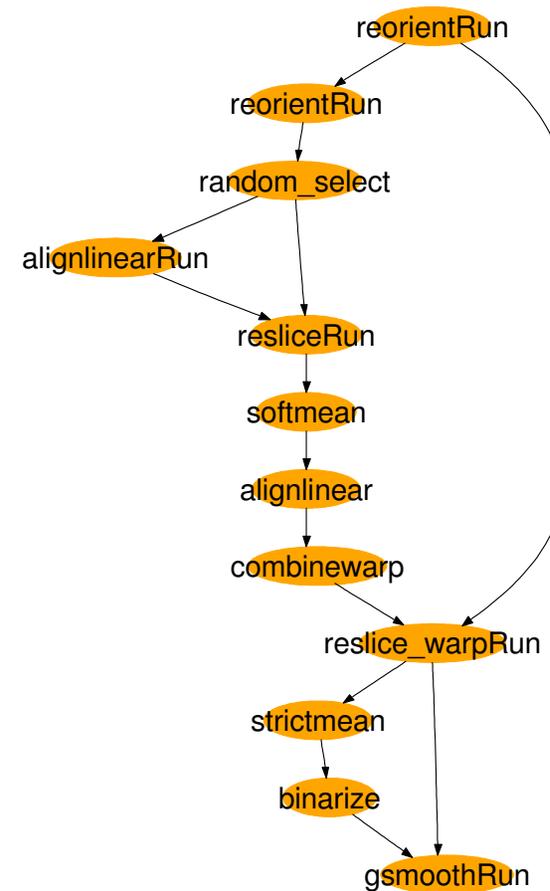
```
(Run snr) functional ( Run r, NormAnat a,  
                      Air shrink ) {  
  Run yroRun = reorientRun( r , "y" );  
  Run roRun = reorientRun( yroRun , "x" );  
  Volume std = roRun[0];  
  Run rndr = random_select( roRun, 0.1 );  
  AirVector rndAirVec = align_linearRun( rndr, std, 12, 1000, 1000, "81 3 3" );  
  Run reslicedRndr = resliceRun( rndr, rndAirVec, "o", "k" );  
  Volume meanRand = softmean( reslicedRndr, "y", "null" );  
  Air mnQAAir = alignlinear( a.nHires, meanRand, 6, 1000, 4, "81 3 3" );  
  Warp boldNormWarp = combinewarp( shrink, a.aWarp, mnQAAir );  
  Run nr = reslice_warp_run( boldNormWarp, roRun );  
  Volume meanAll = strictmean( nr, "y", "null" )  
  Volume boldMask = binarize( meanAll, "y" );  
  snr = gsmoothRun( nr, boldMask, "6 6 6" );
```

```
(Run or) reorientRun (Run ir,  
                      string direction) {  
  foreach Volume iv, i in ir.v {  
    or.v[i] = reorient(iv, direction);  
  }  
}
```

Expressiveness

Lines of code with different encodings

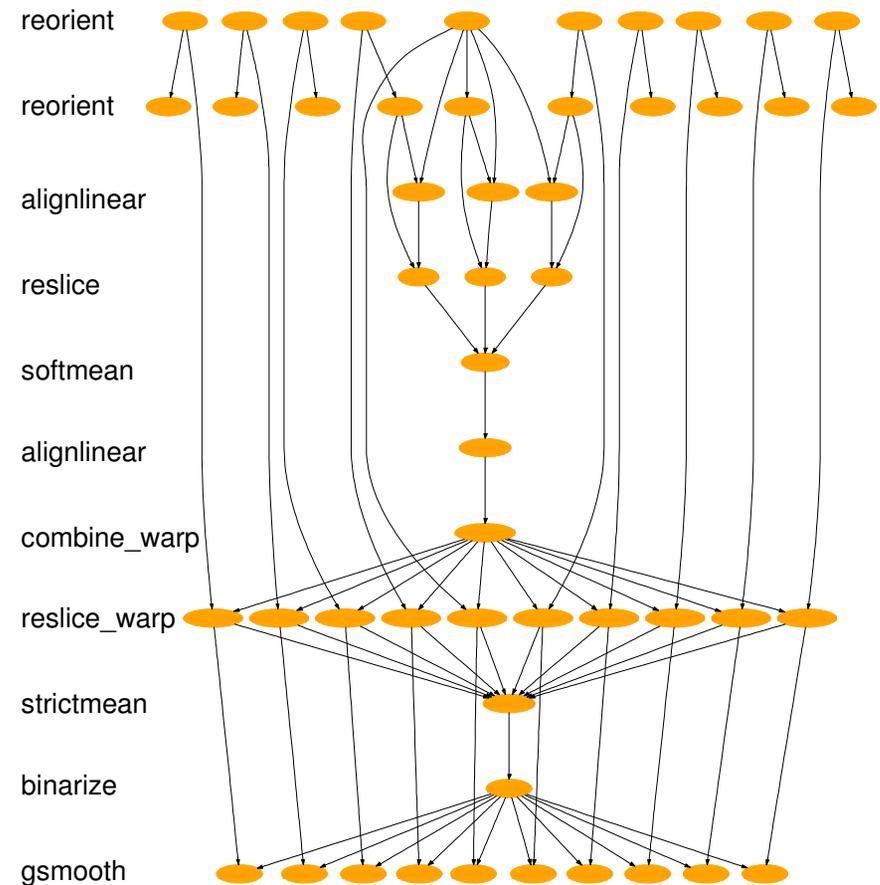
Appln	Script	Generator	Swift Script
ATLAS1	49	72	6
ATLAS2	97	135	10
FILM1	63	134	17
FEAT	84	191	13
AIRSN	215	~400	34



Expressiveness

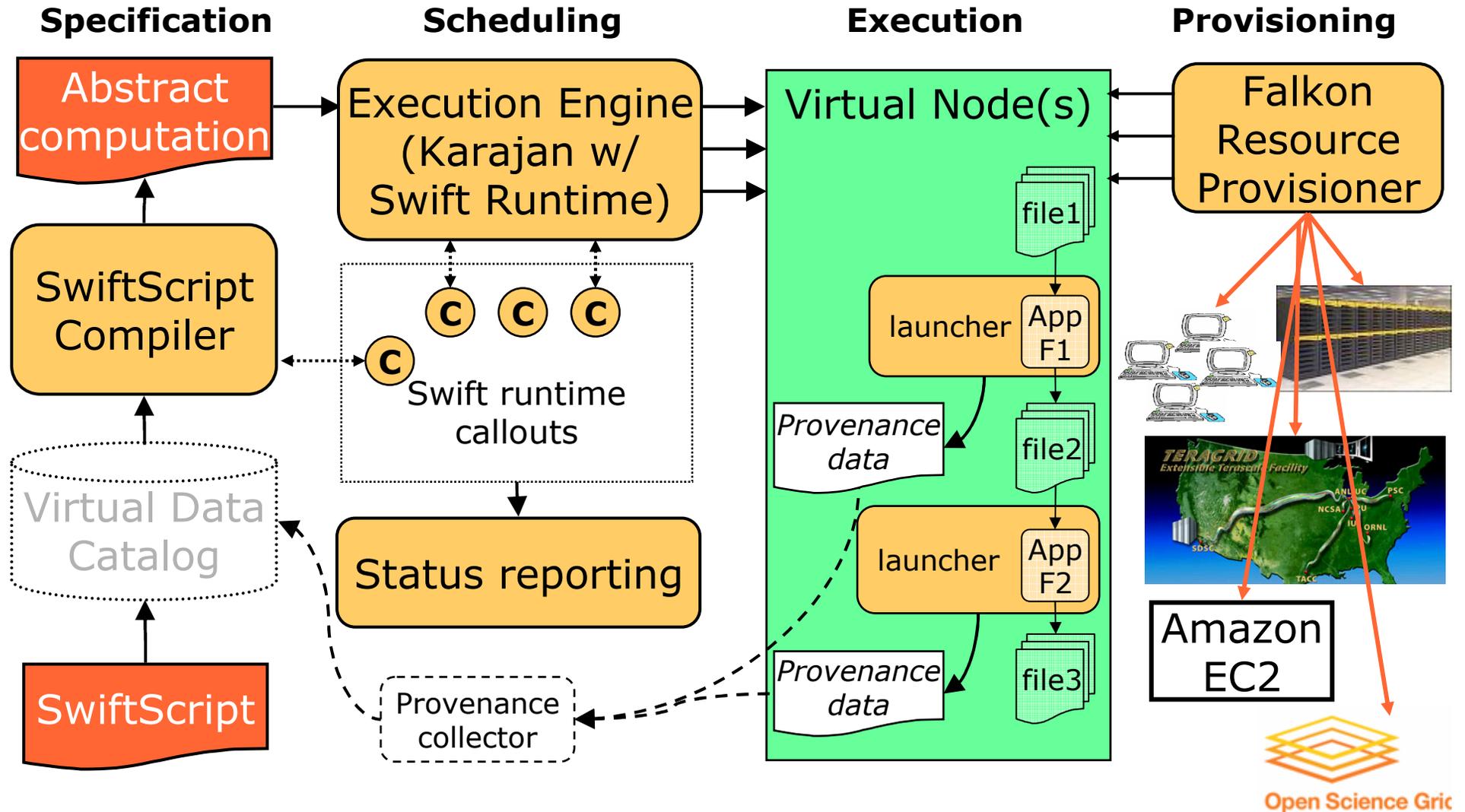
Lines of code with different encodings

Appln	Script	Generator	Swift Script
ATLAS1	49	72	6
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FEAT	84	191	13
AIRSN	215	~400	34





Dynamic Provisioning: Swift Architecture



Yong Zhao, Mihael Hatigan, Ioan Raicu, Mike Wilde, Ben Clifford



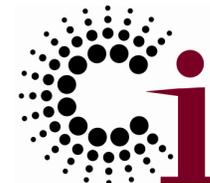
Swift Runtime System

- Runtime system for SwiftScript [SSDBM02,CIDR03,Springer06]
 - ◆ Populate, update, query virtual data products
 - ◆ Schedule, monitor, execute resulting computation on distributed Grid resources
 - ◆ Annotate virtual data products with customized metadata
 - ◆ Trace provenance of virtual data products
- Grid scheduling and optimization
 - ◆ Lightweight execution engine: Karajan
 - ◆ Dynamic resource provisioning
 - ◆ Site selection, data movement, caching
 - ◆ Pipelining, clustering, load balancing
 - ◆ Fault tolerance, exception handling

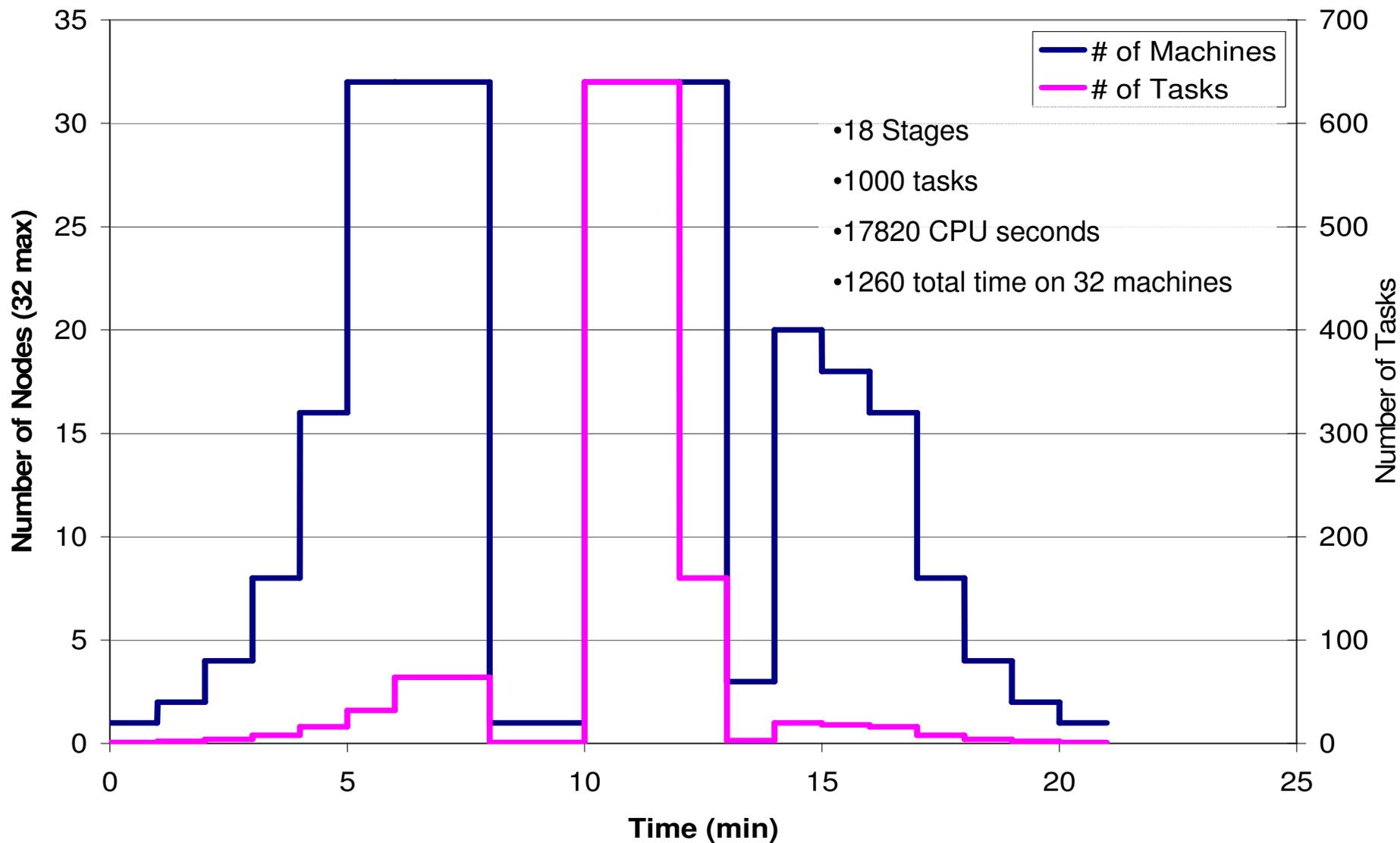


Swift uses Karajan Workflow Engine

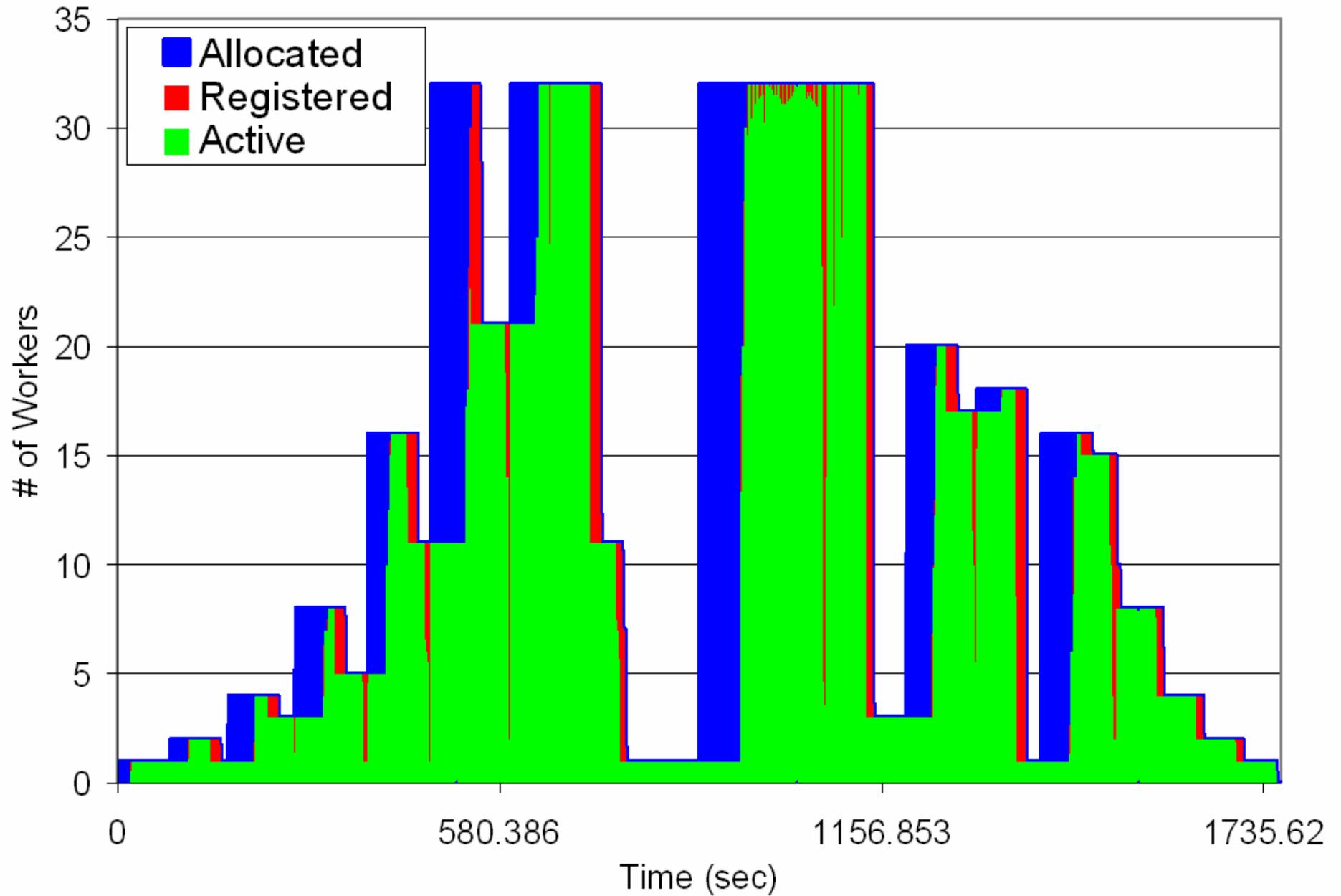
- Fast, scalable threading model
- Suitable constructs for control flow
- Flexible task dependency model
 - ◆ “Futures” enable pipelining
- Flexible provider model allows for use of different run time environments
 - ◆ Job execution and data transfer
 - ◆ Flow controlled to avoid resource overload
- Workflow client runs from a Java container



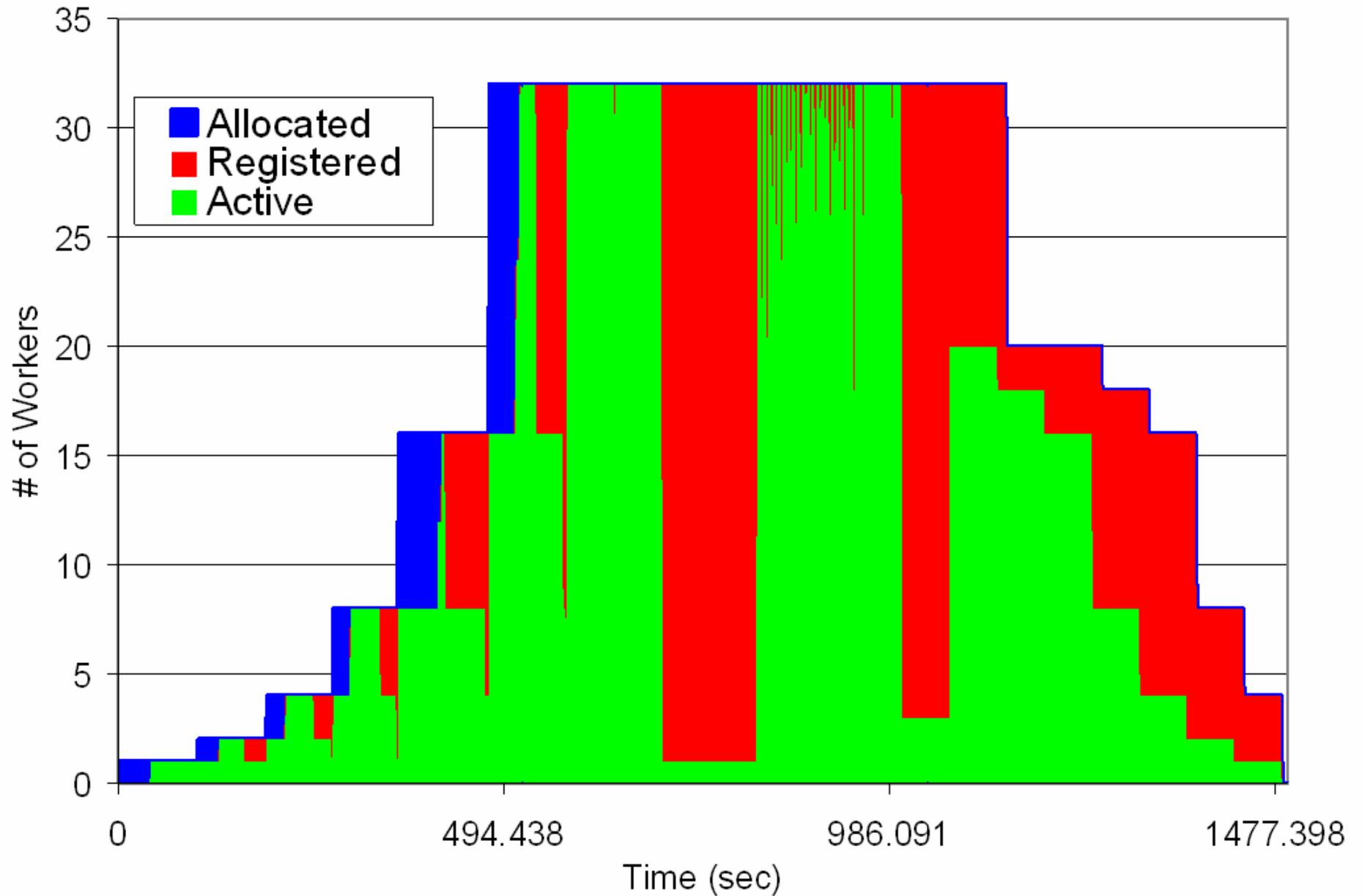
Synthetic Benchmark



Release after 15 Seconds Idle

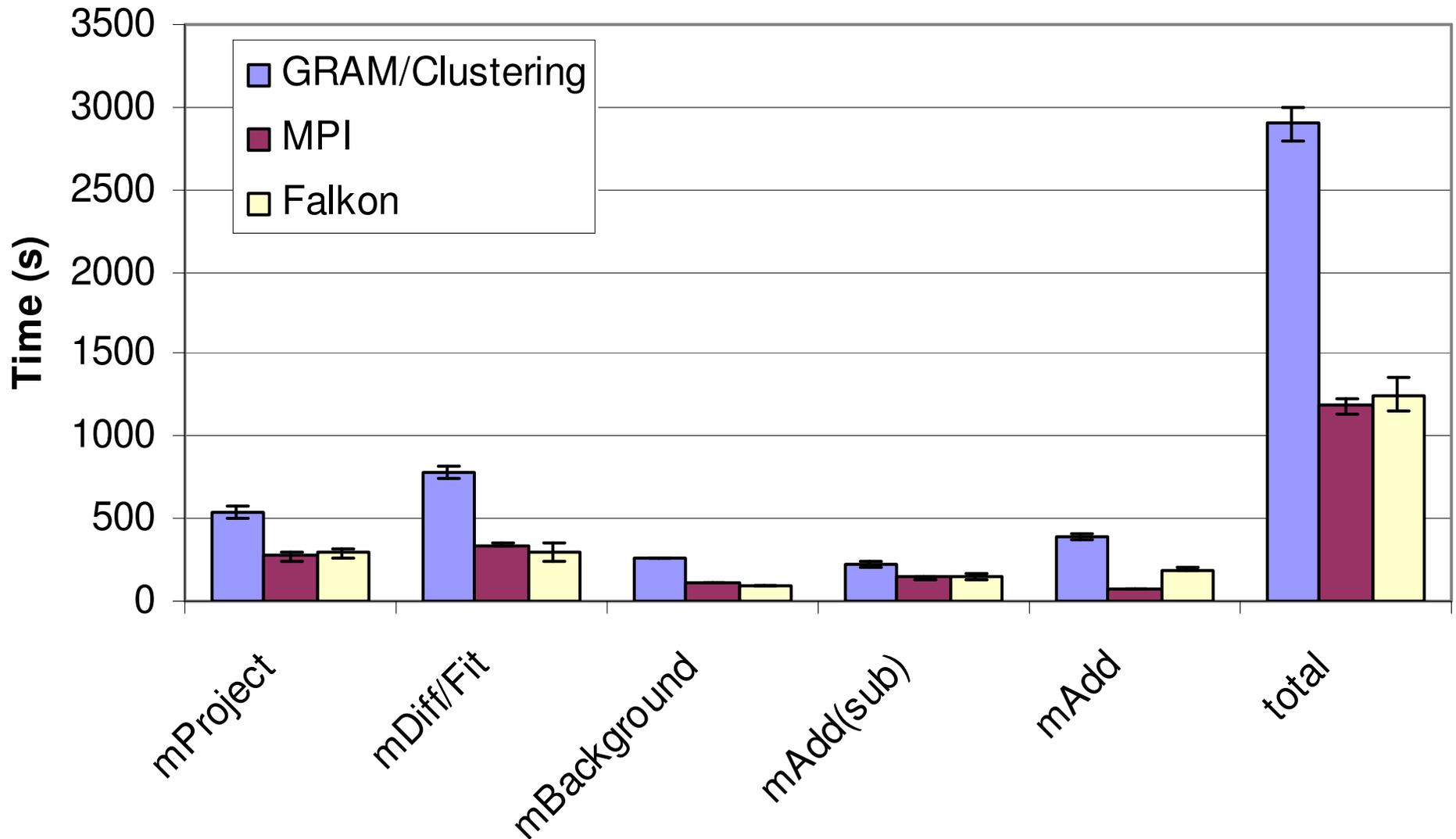


Release after 180 Seconds Idle

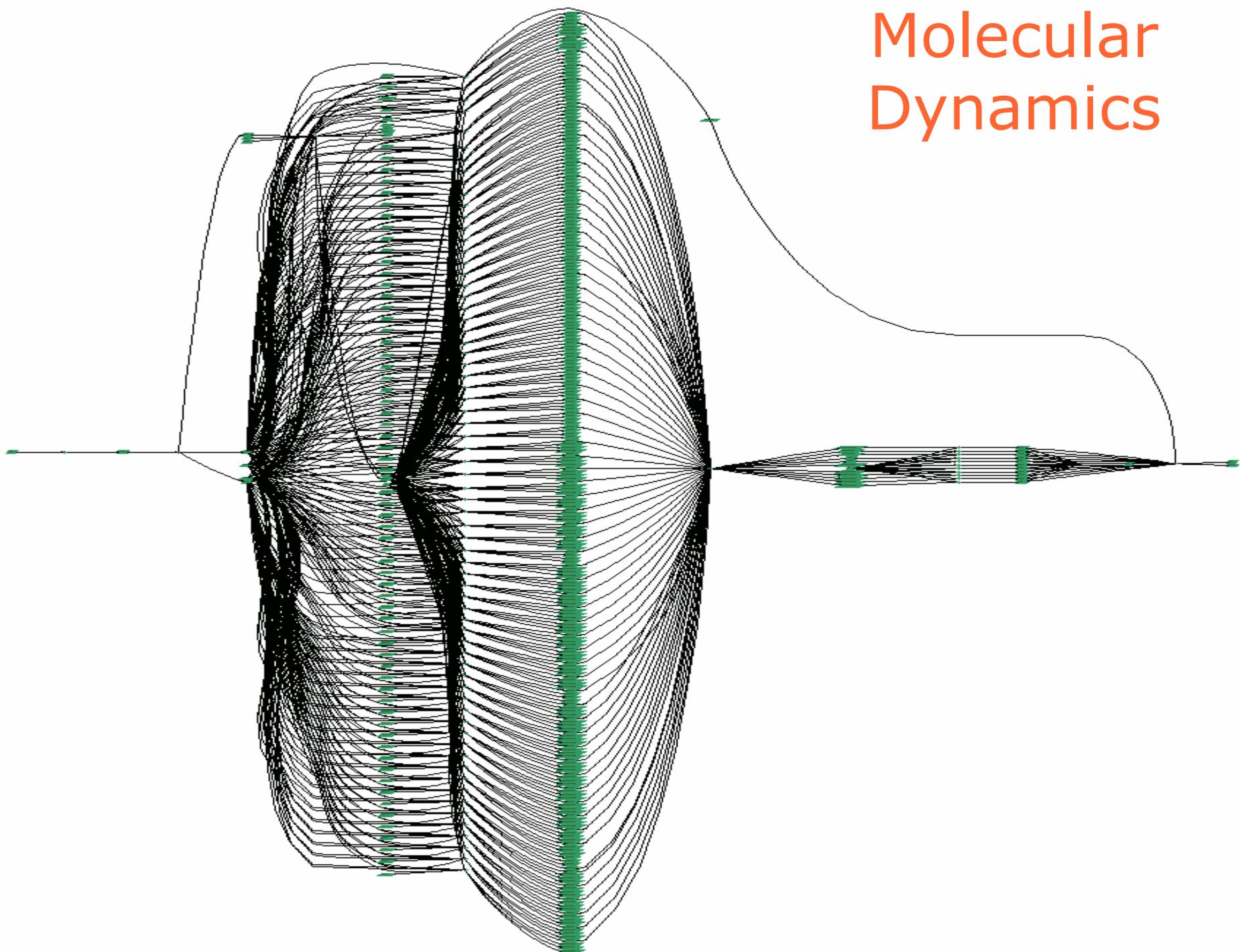


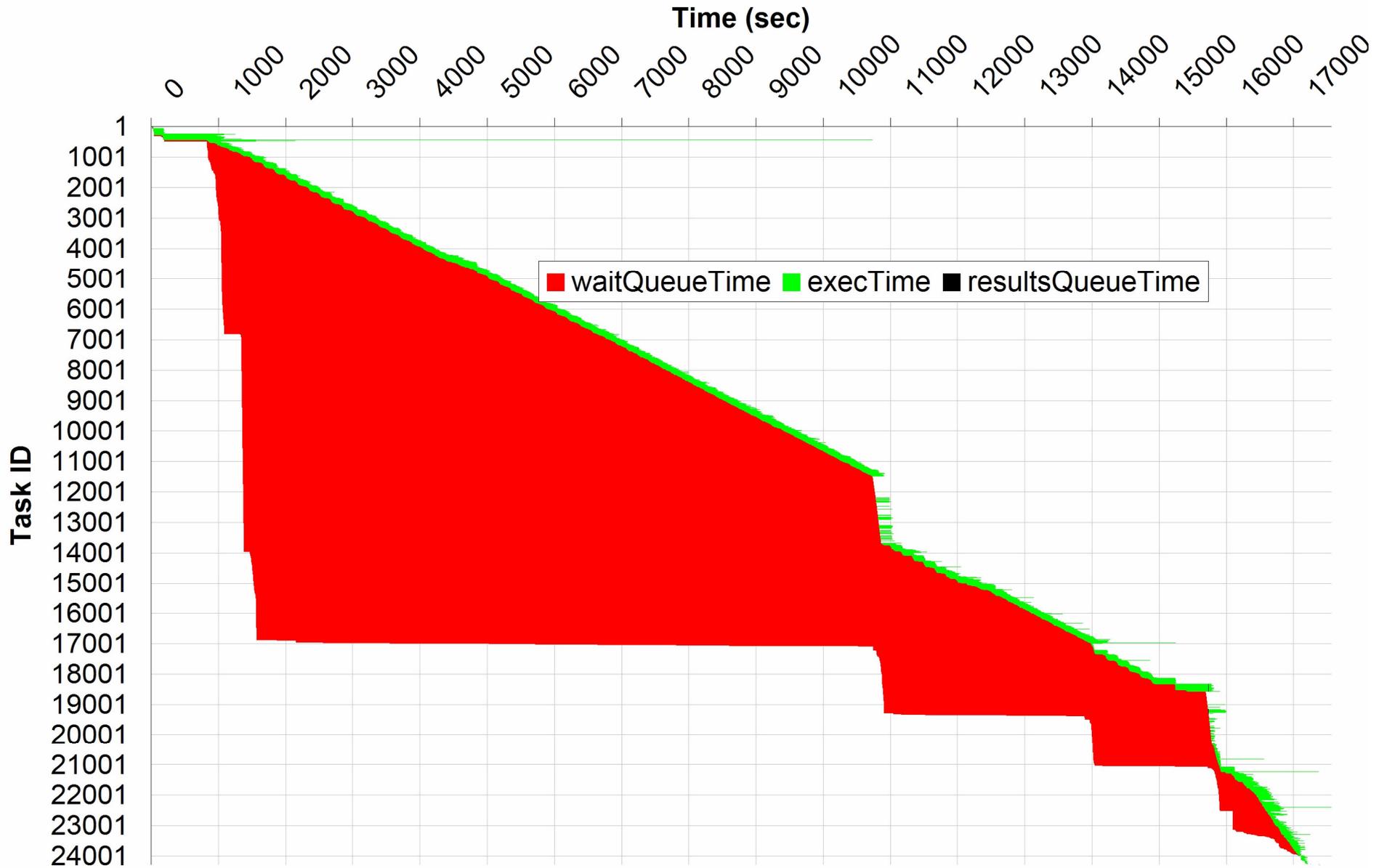


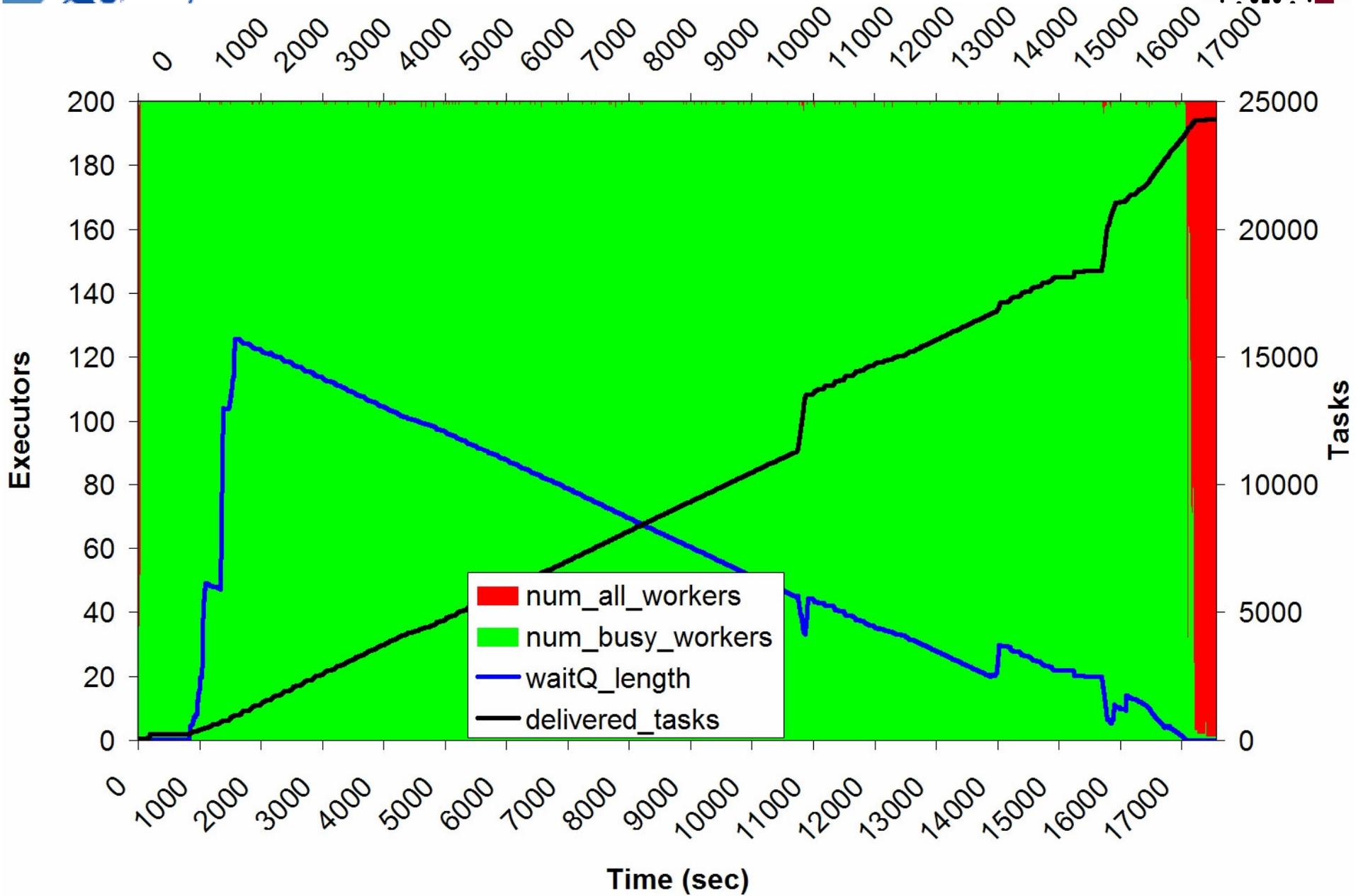
Montage



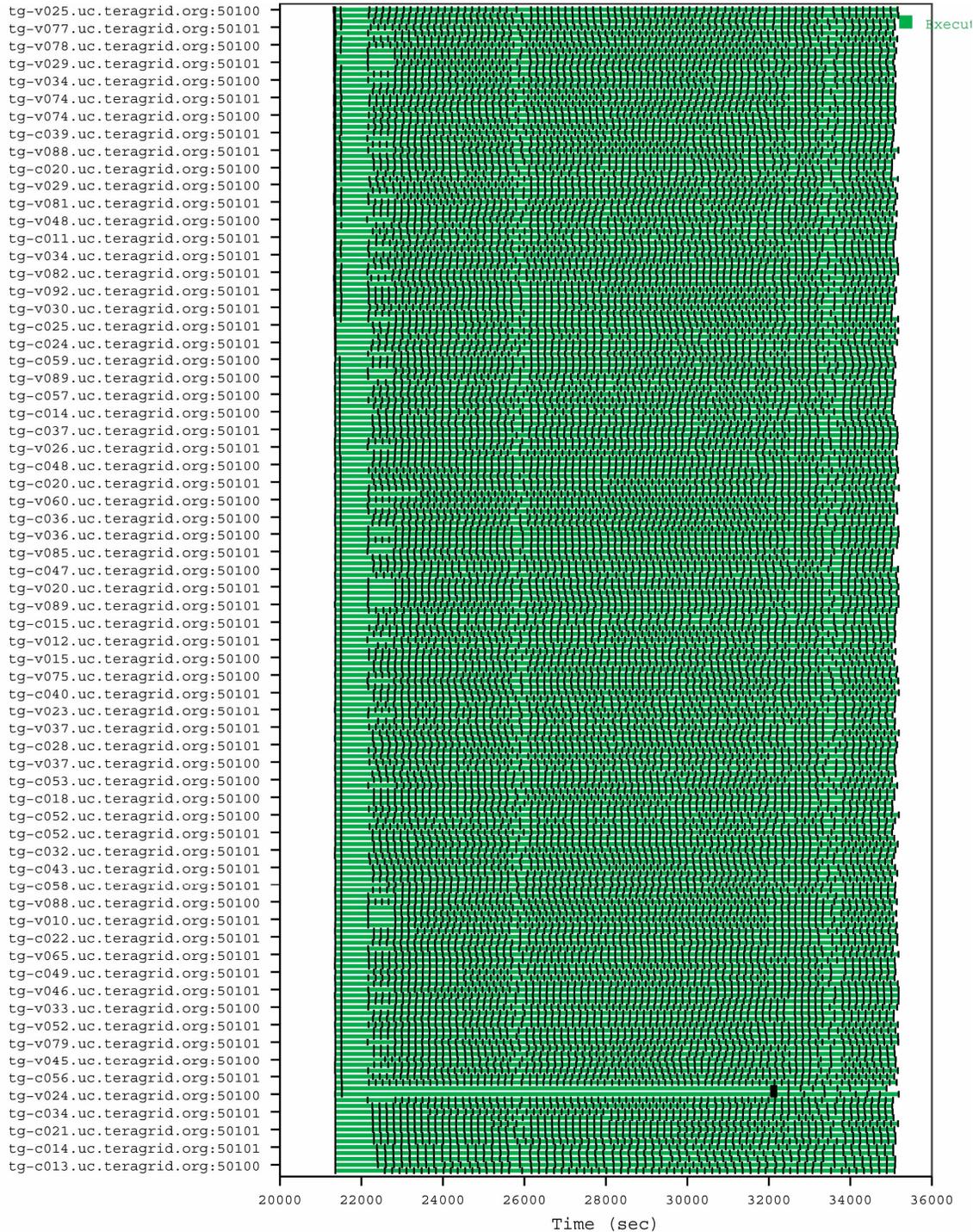
Molecular Dynamics

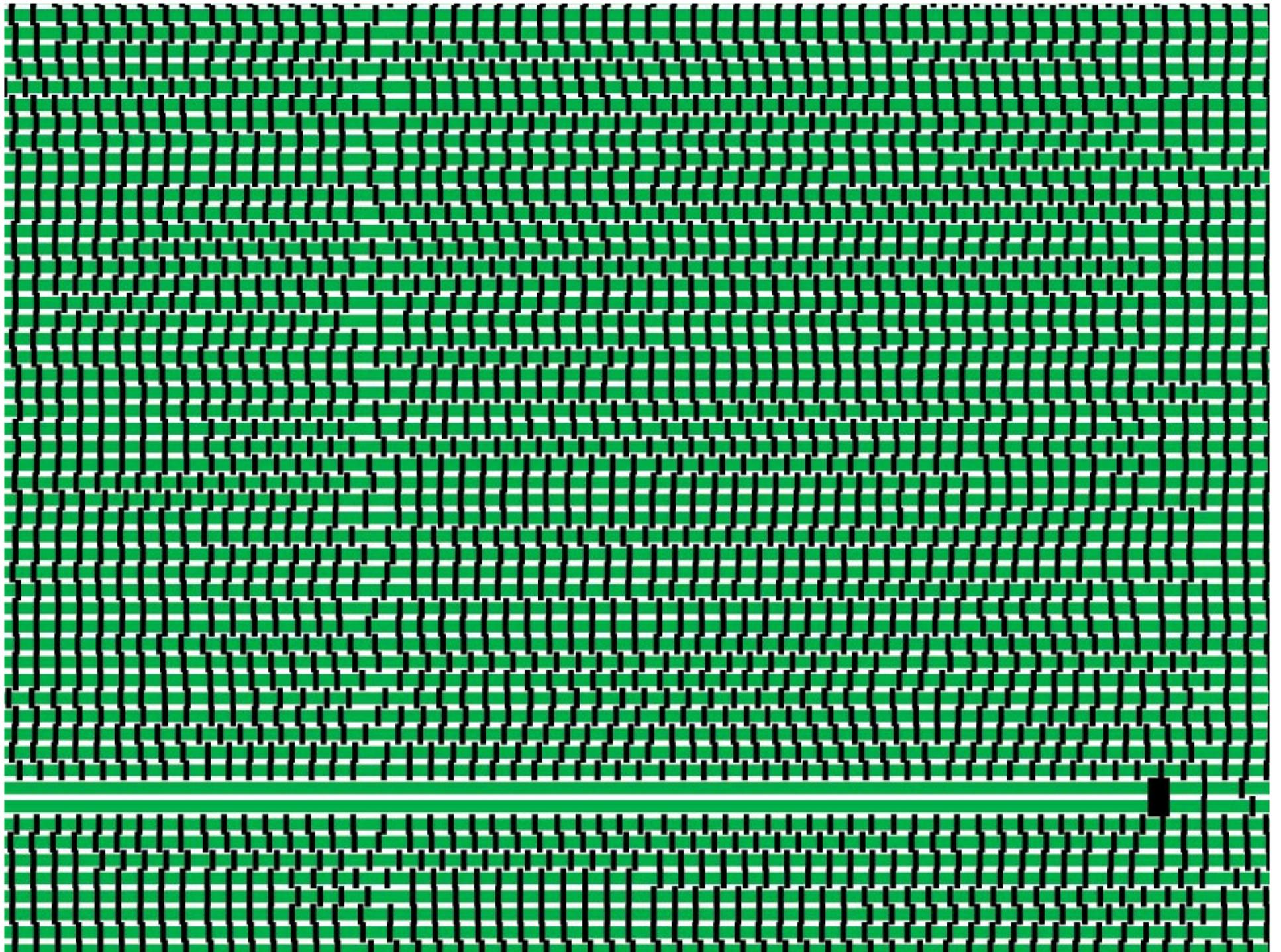


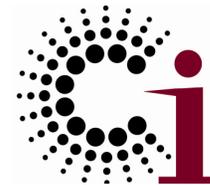




Executors



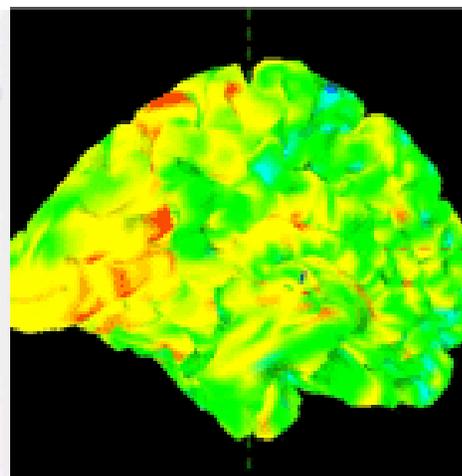
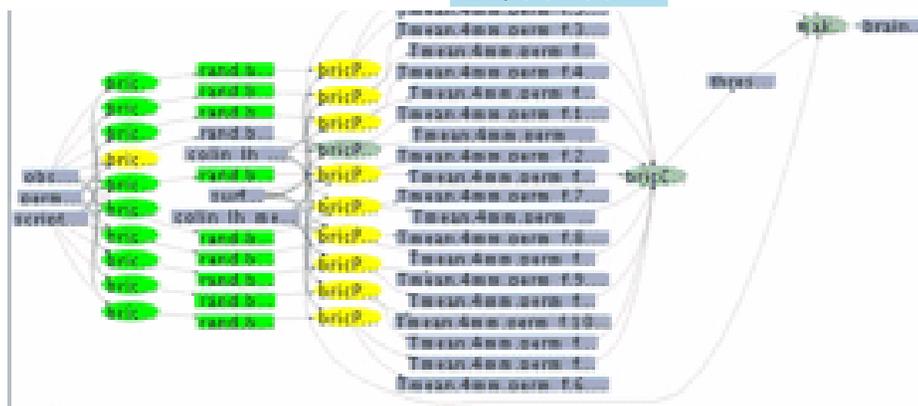




Application Example: ACTIVAL: Neural Activation Validation



Identifies clusters of neural activity not likely to be active by random chance: switch labels of the conditions for one or more participants; calculate the delta values in each voxel, re-calculate the reliability of delta in each voxel, and evaluate clusters found. If the clusters in data are greater than the majority of the clusters found in the permutations, then the null hypothesis is refuted indicating that clusters of activity found in our experiment are not likely to be found by chance.



Work by S. Small,
U. Hasson, UChicago

SwiftScript Program ACTIVAL – Datatypes & Utilities

```
type script {}
type brainMeasurements{}
type precomputedPermutations{}
type brainClusterTable {}
type brainDatasets{ brainDataset b[]; }
type brainClusters{ brainClusterTable c[]; }

type fullBrainData {}
type fullBrainSpecs {}
type brainDataset {}
```

// Procedure to run "R" statistical package

```
(brainDataset t) bricRInvoke (script permutationScript, int iterationNo,
    brainMeasurements dataAll, precomputedPermutations dataPerm) {
    app { bricRInvoke @filename(permutationScript) iterationNo
        @filename(dataAll) @filename(dataPerm); }
}
```

// Procedure to run AFNI Clustering tool

```
(brainClusterTable v, brainDataset t) bricCluster (script clusterScript,
    int iterationNo, brainDataset randBrain, fullBrainData brainFile,
    fullBrainSpecs specFile) {
    app { bricPerlCluster @filename(clusterScript) iterationNo
        @filename(randBrain) @filename(brainFile)
        @filename(specFile); }
}
```

// Procedure to merge results based on statistical likelihoods

```
(brainClusterTable t) bricCentralize ( brainClusterTable bc[]) {
    app { bricCentralize @filenames(bc); }
}
```

ACTIVAL: Dataset Iteration Procedures

// Procedure to iterate over the data collection

```
(brainClusters randCluster, brainDatasets dsetReturn) brain_cluster  
  (fullBrainData brainFile, fullBrainSpecs specFile)  
{  
  int sequence[]={1:2000};  
  
  brainMeasurements      dataAll<fixed_mapper; file="obs.imit.all">;  
  precomputedPermutations dataPerm<fixed_mapper; file="perm.matrix.11">;  
  script                  randScript<fixed_mapper; file="script.obs.imit.tibi">;  
  script                  clusterScript<fixed_mapper; file="surfclust.tibi">;  
  brainDatasets           randBrains<simple_mapper; prefix="rand.brain.set">;  
  
  foreach int i in sequence {  
    randBrains.b[i] = bricRInvoke(randScript,i,dataAll,dataPerm);  
    brainDataset rBrain=randBrains.b[i];  
    (randCluster.c[i],dsetReturn.b[i]) =  
      bricCluster(clusterScript,i,rBrain, brainFile,specFile);  
  }  
}
```

ACTIVAL: Main Program

// Declare datasets

```
fullBrainData      brainFile<fixed_mapper; file="colin_lh_mesh140_std.pial.asc">;
fullBrainSpecs    specFile<fixed_mapper; file="colin_lh_mesh140_std.spec">;

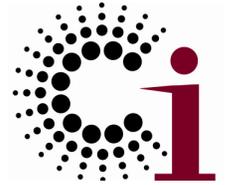
brainDatasets     randBrain<simple_mapper; prefix="rand.brain.set">;
brainClusters     randCluster<simple_mapper; prefix="Tmean.4mm.perm",
                  suffix="_ClstTable_r4.1_a2.0.1D">;
brainDatasets     dsetReturn<simple_mapper; prefix="Tmean.4mm.perm",
                  suffix="_Clustered_r4.1_a2.0.niml.dset">;
brainClusterTable clusterThresholdsTable<fixed_mapper; file="thresholds.table">;
brainDataset      brainResult<fixed_mapper; file="brain.final.dset">;
brainDataset      origBrain<fixed_mapper; file="brain.permutation.1">;
```

// Main program – executes the entire application

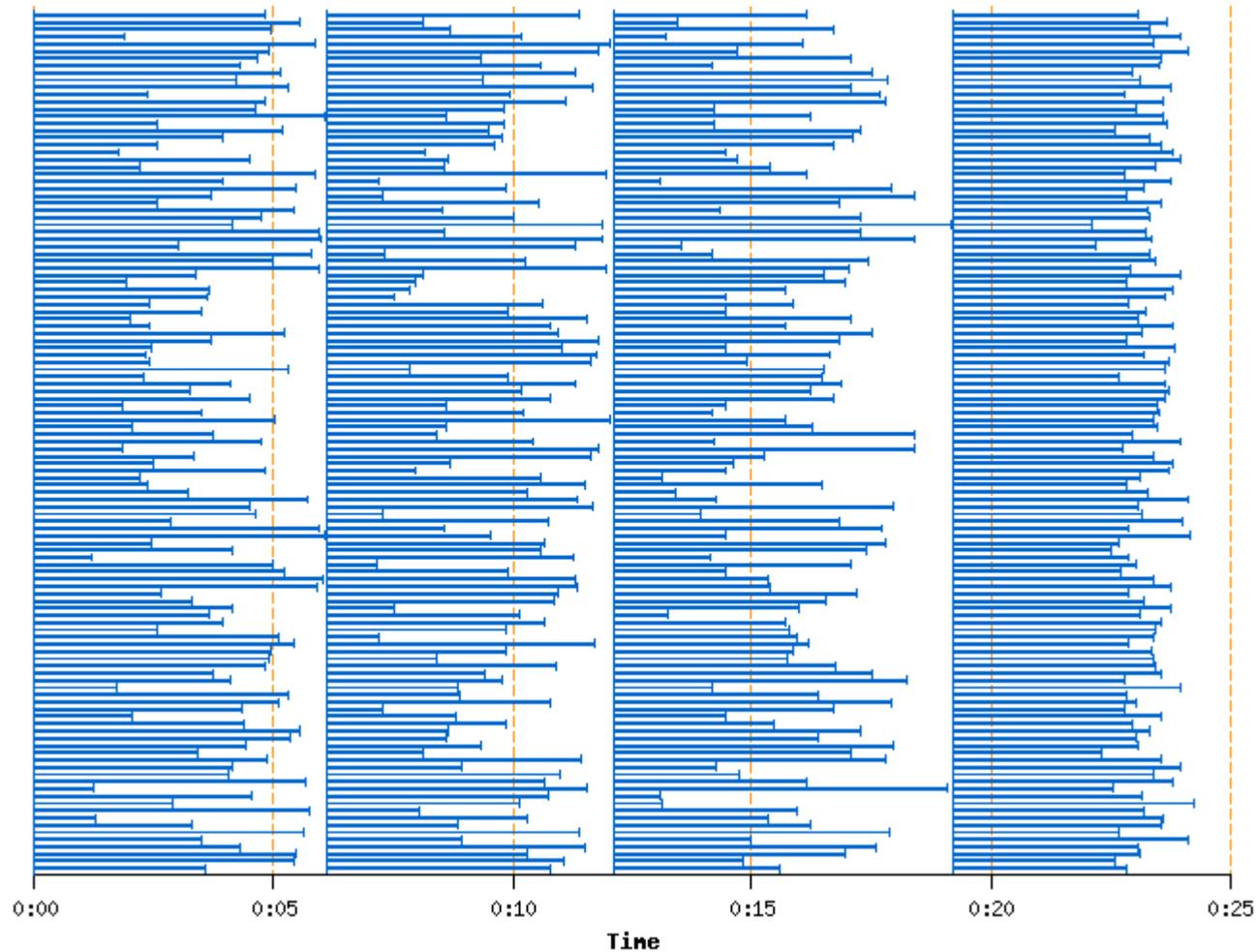
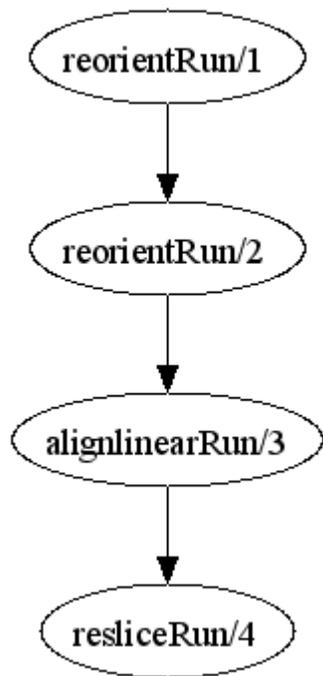
```
(randCluster, dsetReturn) = brain_cluster(brainFile, specFile);

clusterThresholdsTable = bricCentralize (randCluster.c);

brainResult = makebrain(origBrain,clusterThresholdsTable,brainFile,specFile);
```

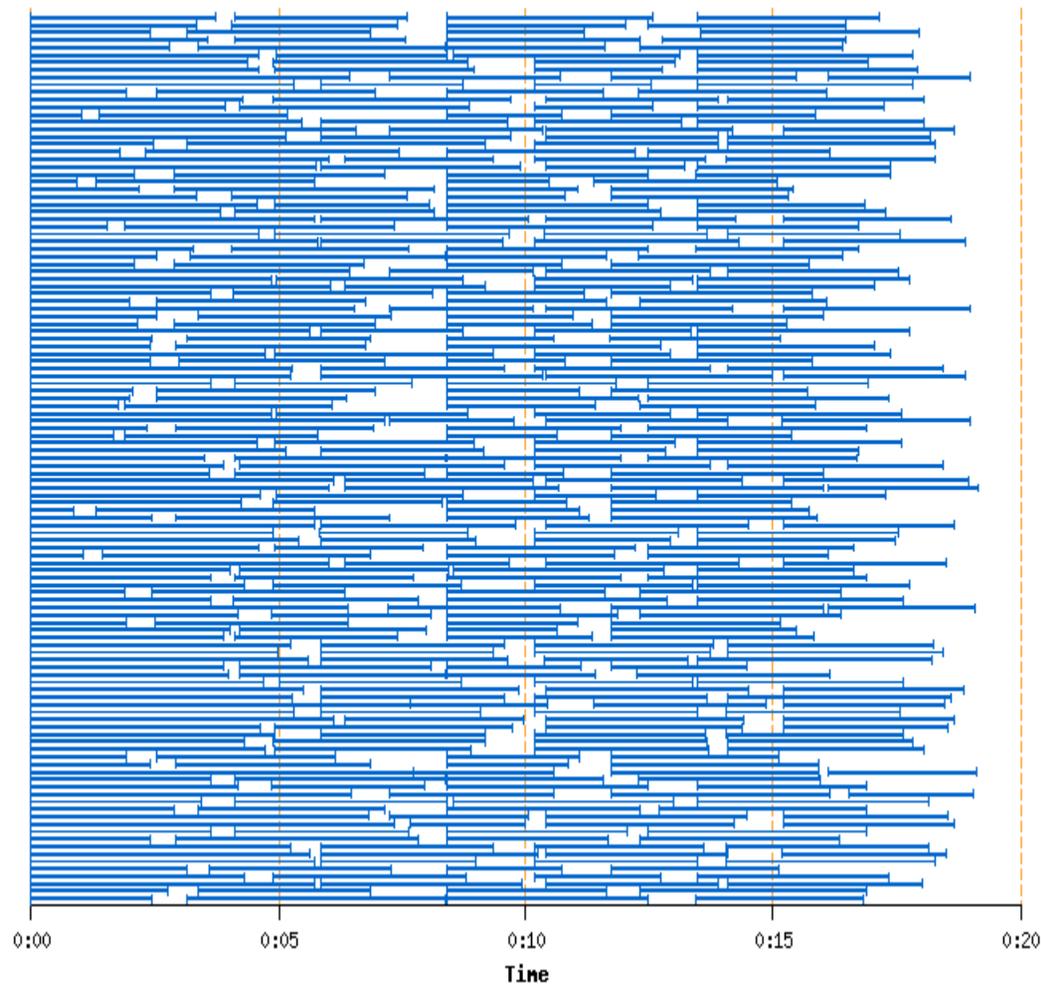
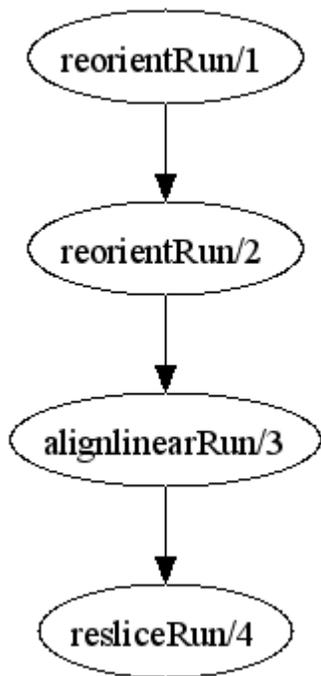


Example Performance Optimizations





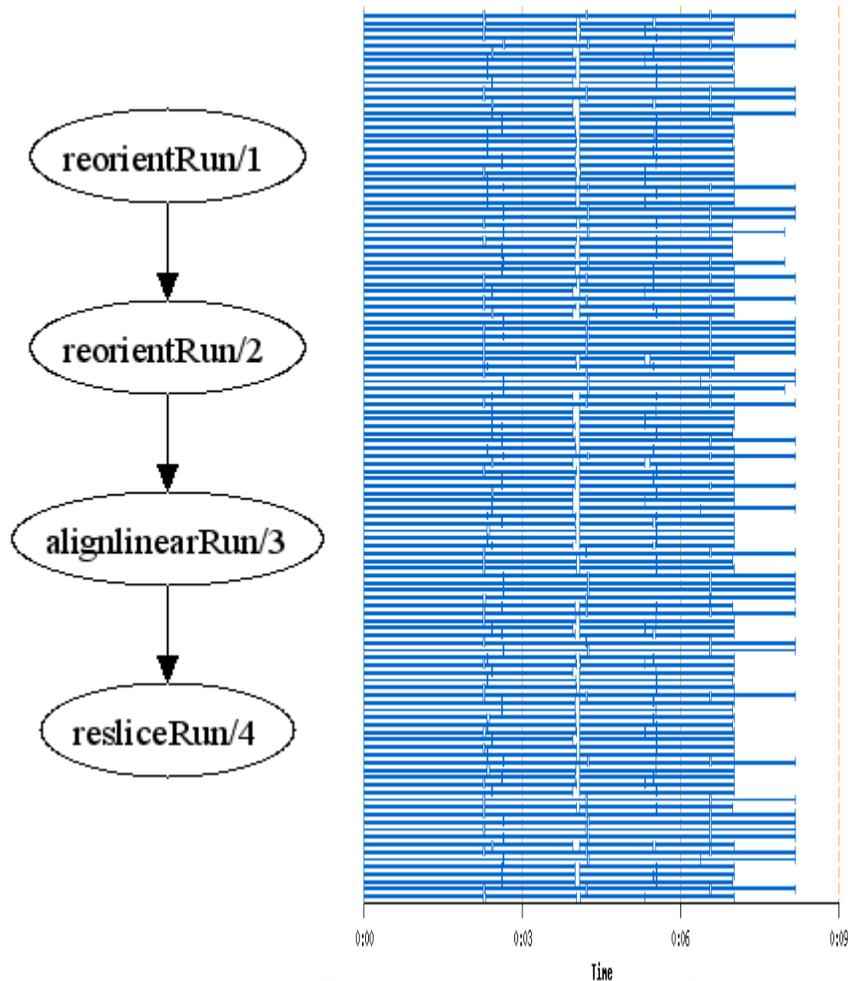
Example Performance Optimizations



Pipelining



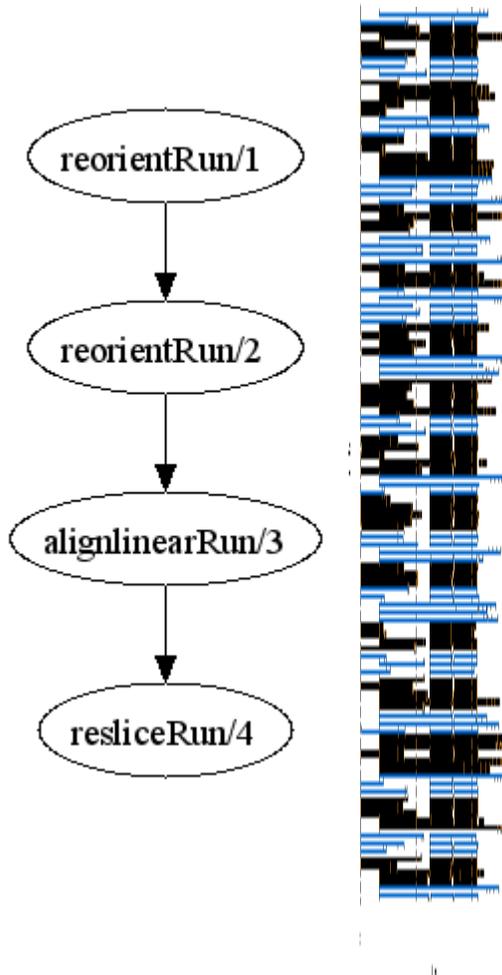
Example Performance Optimizations



Pipelining + **clustering**



Example Performance Optimizations



Pipelining + provisioning



Other Applications



Application	#Jobs/computation	Levels
ATLAS* HEP Event Simulation	500K	1
fMRI DBIC* AIRSN Image Processing	100s	12
FOAM Ocean/Atmosphere Model	2000 (core app runs 250 8-CPU jobs)	3
GADU* Genomics: (14 million seq. analyzed)	40K	4
HNL fMRI Aphasia Study	500	4
NVO/NASA* Photorealistic Montage/Morphology	1000s	16
QuarkNet/I2U2* Physics Science Education	10s	3-6
RadCAD* Radiology Classifier Training	1000s	5
SIDGrid EEG Wavelet Proc, Gaze Analysis, ...	100s	20
SDSS* Coadd, Cluster Search	40K, 500K	2, 8



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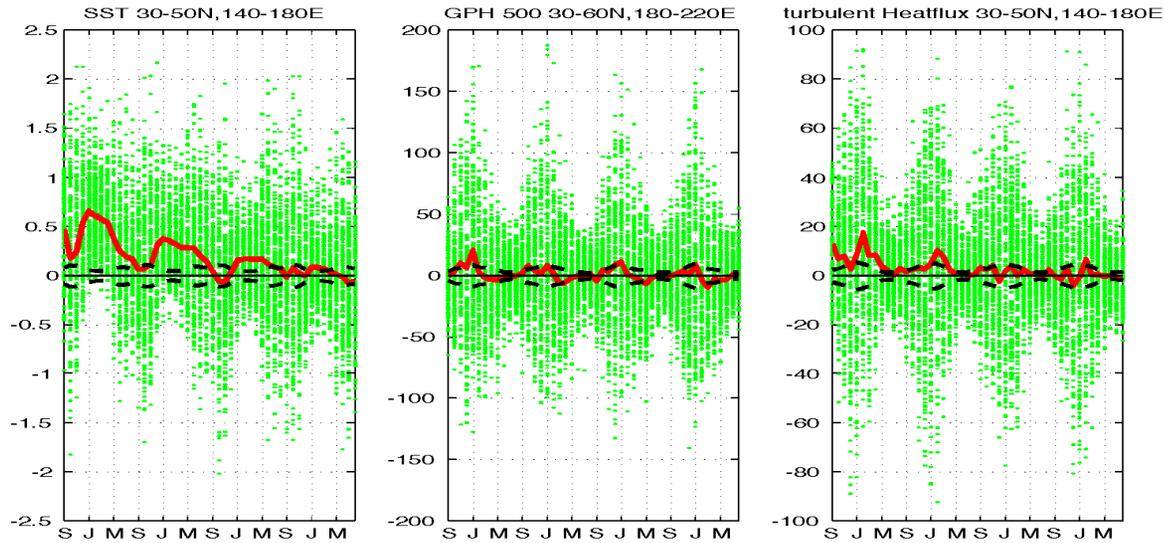
Fast Ocean Atmosphere Model



160 ensemble members = 2.5 months to run

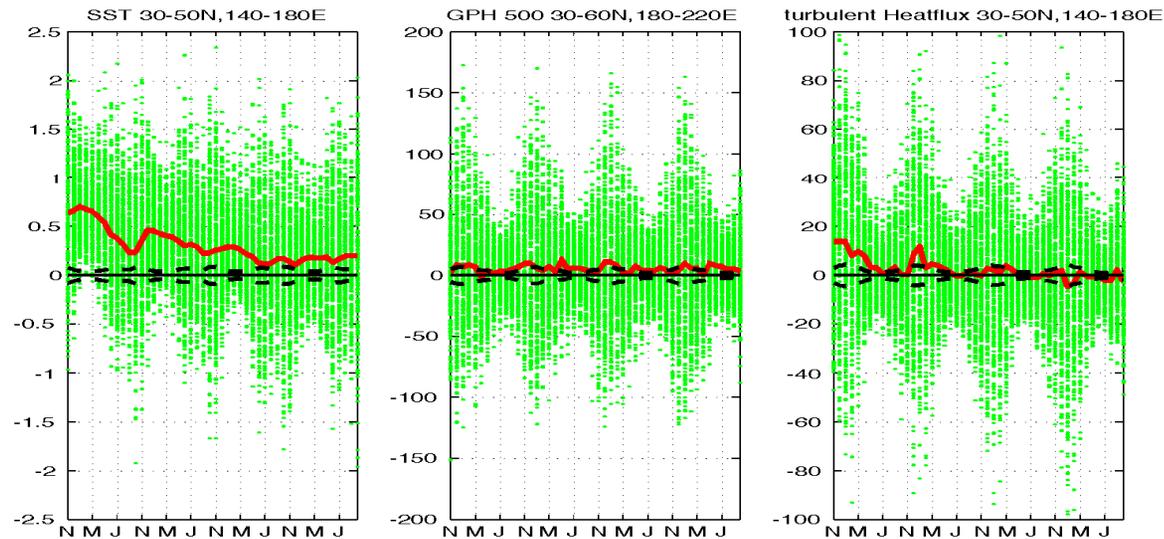
NCAR

*Manual config,
execution,
bookkeeping*



250 ensemble members = 4 days to run

*VDS on Teragrid
Automated*



Green: each ensemble Red: ensemble mean

*Visualization
courtesy Pat
Behling and
Yun Liu, UW
Madison*



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Swift System



www.ci.uchicago.edu/swift

- Clean separation of logical/physical concerns
 - ◆ **XDTM** specification of logical data structures
- + Concise specification of parallel programs
 - ◆ **SwiftScript**, with iteration, etc.
- + Efficient execution on distributed resources
 - ◆ Lightweight threading, dynamic provisioning, Grid interfaces, pipelining, load balancing
- + Rigorous provenance tracking and query
 - ◆ Virtual data schema & automated recording
- **Improved usability and productivity**
 - ◆ Demonstrated in numerous applications